

The Long-term Decline of the U.S. Job Ladder

Aniket Baksy

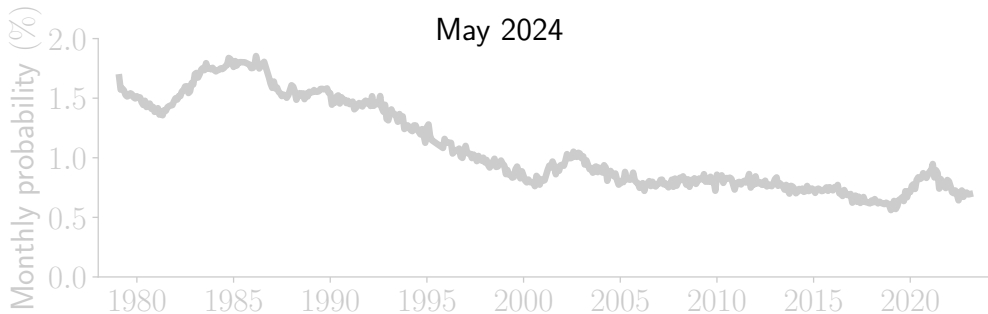
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- EE mobility is integral for:
 1. Micro: life-cycle wage growth (Topel and Ward, '92)
 2. Macro: alleviating misallocation (Moscarini and Postel-Vinay '17, Bilal et al. '22)
- Yet little is known about long run trends in EE mobility in the U.S.
 - * No data before 1994, data post 1994 have issues

This paper

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 - * Prototypical partial equilibrium job-ladder model
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 1. Are workers better matched today?
 2. Decline in matching efficiency?
 3. Increased labor market concentration?

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- Evaluates 3 hypotheses behind this decline
 1. Are workers better matched today? No
 2. Decline in matching efficiency? No
 3. Increased labor market concentration? May account for 40% of decline

Roadmap

- Theory
- Data and validation
- Three facts on EE mobility
- Explanations for the decline
- Conclusion

Theory

Accounting framework

- We observe 10 workers earning less than w at t and 8 at $t + 1$

10
⏟
earning $\leq w$ at t

8
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earning $\leq w$ at $t + 1$

Accounting framework

- We observe 10 workers earning less than w at t and 8 at $t + 1$
- Between t and $t + 1$:
 - * 1 of these workers separates into nonemployment
 - * 2 nonemployed find jobs that pay less than w

$$\underbrace{10}_{\text{earning} \leq w \text{ at } t} - \underbrace{1}_{\text{outflows}} + \underbrace{2}_{\text{inflows}}$$

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Caveat: this captures transitions toward higher-paying jobs

A partial equilibrium job ladder model

- Unit mass of risk-neutral, infinitely lived workers
- Mass $1 - e_t$ of **nonemployed**:
 - * get job offer with prob. λ_t^n
 - * draw from *exogenous* wage offer cdf $F_{t+1}^n(w)$
 - * assume all offers are accepted
- Mass e_t of **employed**:
 - * lose job with prob. δ_t
 - * get job offer with prob. λ_t^e
 - * draw from *exogenous* wage offer distribution $F_{t+1}^e(w)$
 - * **only accept offers that pay a higher wage**

Labor market flows

- Share of workers earning wage w , $g_t(w)$ evolves according to

$$g_{t+1}(w) e_{t+1} - g_t(w) e_t =$$

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Deriving EE mobility

- Integrating (1) and rearranging

$$\underbrace{\lambda_t^e (1 - F_{t+1}^e(w))}_{\equiv sep_t^e(w)} = 1 - \underbrace{\frac{G_{t+1}(w)}{G_t(w)} \frac{e_{t+1}}{e_t}}_{\text{change in emp.}} + \underbrace{\lambda_t^n \frac{F_{t+1}^n(w)}{G_t(w)} \frac{1 - e_t}{e_t}}_{\text{hires from nonemp.}} - \underbrace{\delta_t}_{\text{sep. to nonemp.}}$$

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- And so EE mobility is

$$EE_t \equiv \underbrace{\lambda_t^e}_{\text{job-finding prob.}} \underbrace{\int_{-\infty}^{\infty} \left(1 - F_{t+1}^e(w)\right) dG_t(w)}_{\text{average acceptance prob.}} = \underbrace{\int_{-\infty}^{\infty} \text{sep}_t^e(w) dG(w)}_{\text{average poaching separation prob.}}$$

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- Recover EE given

* $G_t, G_{t+1}, F_{t+1}^n, e_t, e_{t+1}, \delta_t, \lambda_t^n$, no need to observe $F_{t+1}^e(w)$

Understanding identification

- In SS, employment in and outflows equal ($\delta_t e_t = \lambda_t^n (1 - e_t)$)
 \Rightarrow separation prob. is $\lambda_t^e \left(1 - F_{t+1}^e(w)\right) = \delta_t \frac{F_{t+1}^n(w) - G_t(w)}{G_t(w)}$

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- EE mobility identified by gap $F_{t+1}^n(w) - G_t(w)$



Data and validation

CPS: 1979-2023

- Rotating panel of 60,000 households
- Employment status in months 1-4 and 13-16
 - * pin down e_t and e_{t+1} , λ_t^n , and δ_t
- Wages recorded *only* in months 4 and 16
 - * project on $\text{age} \times \text{race} \times \text{gender} \times \text{education} \times \text{year}$, $\text{state} \times \text{date}$ fixed effects
 - * bin residuals in 50 bins
 - * pin down $G_t(w)$, $G_{t+1}(w)$, and $F_t^n(w)$ (conditional on previous nonemp.)

Validation with *raw* EE: post 1996

- Construct EE mobility in SIPP between 1996 and 2013

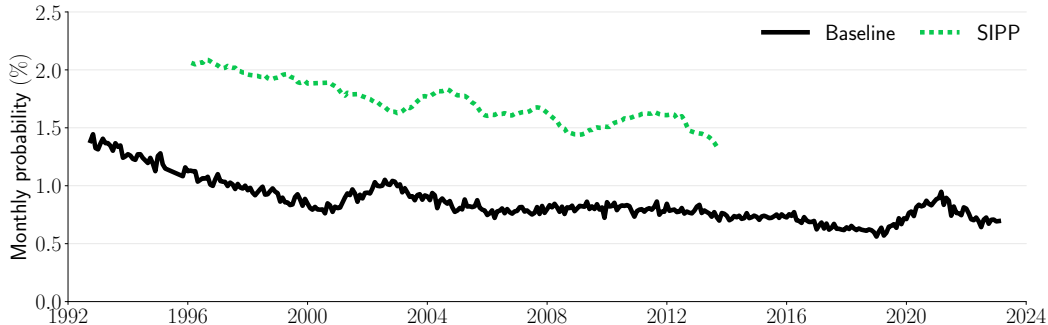
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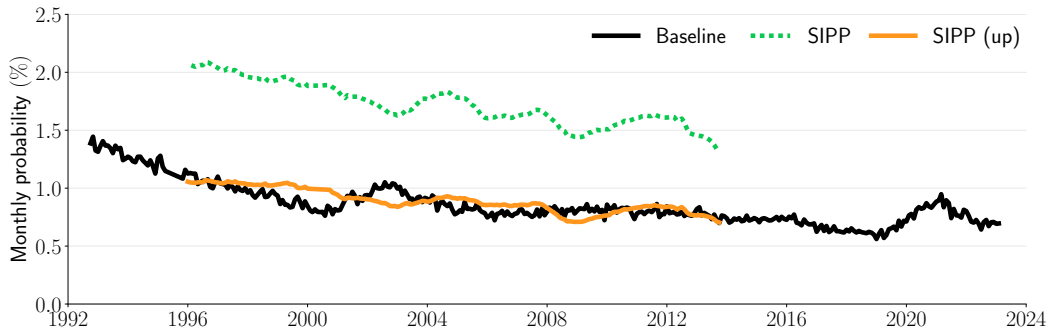
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Validation with *raw* EE: post 1996

- Construct EE mobility in SIPP between 1996 and 2013
- Series matches up well with SIPP (up) in both levels and change



Validation with *raw* EE: pre 1996

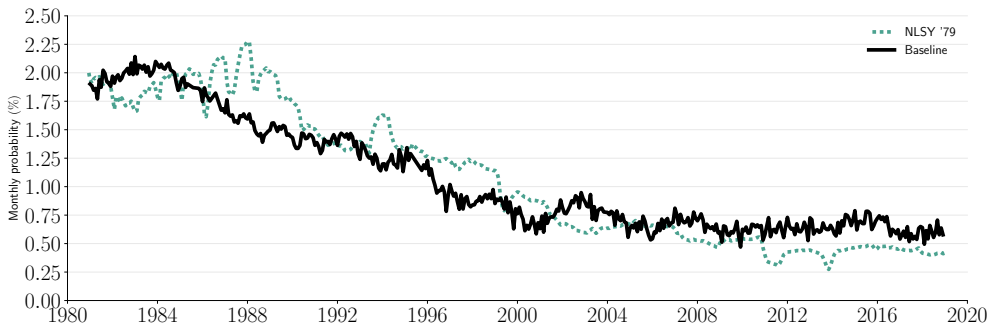
- No data allows to compute overall EE (up) pre 1996

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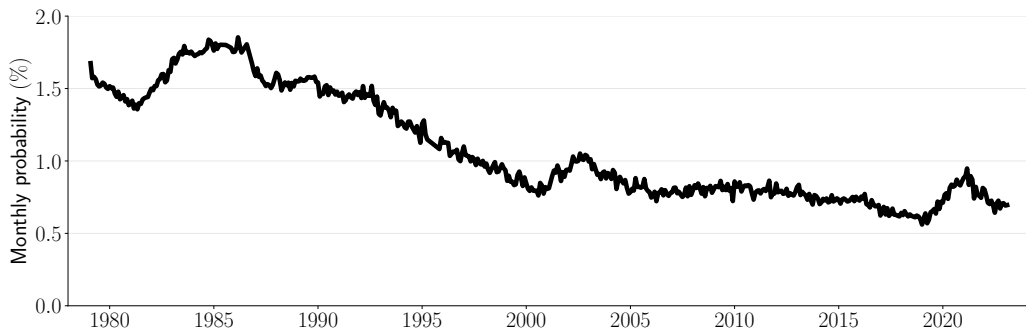


Three facts on EE mobility

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1. EE mobility decline since 1979

Fact 1: EE mobility decline since 1979



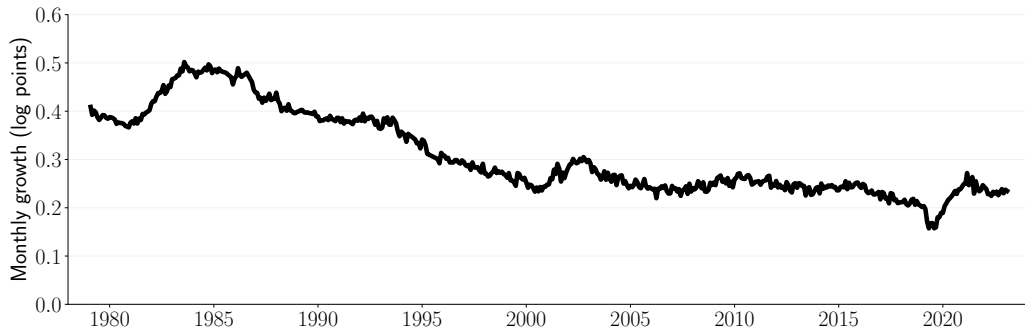
► Alternative specifications

- EE mobility towards higher-paying jobs declined by half from 1979 to 2023

Three facts on EE mobility

1. EE mobility decline since 1979
2. Associated wage growth fell by more than 1 pp.

Fact 2: Associated wage growth fell by more than 1 pp.



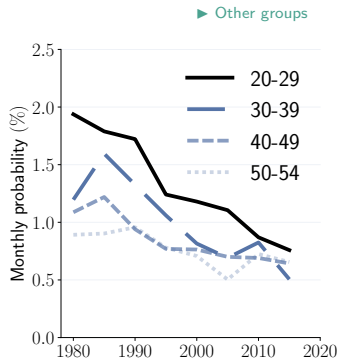
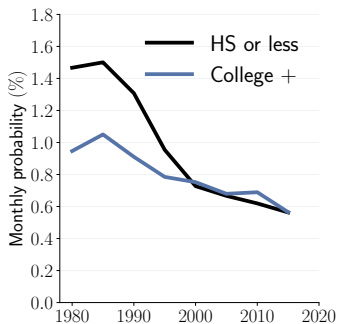
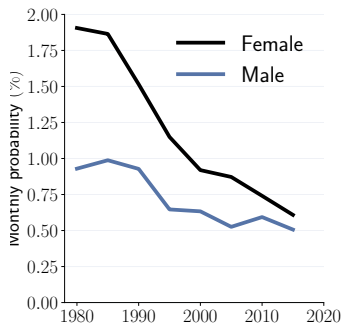
► OTJ growth

- *Annual wage growth* associated with EE mobility fell by more than 1 pp.

Three facts on EE mobility

1. EE mobility decline since 1979
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3. Larger decline for women, less educated, young workers

Fact 3: Larger decline for women, less educated, young



Fact 3: Shift-share exercise

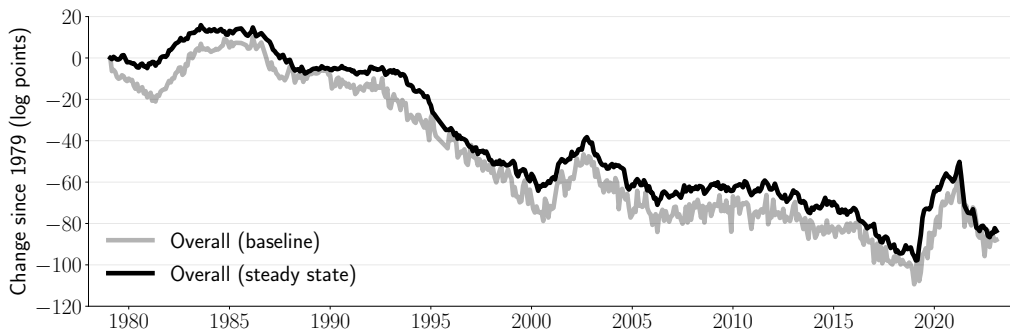
- Age \times education contribute $\sim 1/3$ of overall decline

	Gender	Race	Education	Age	Age \times Education
Composition	-3.3%	-1.6%	11.9%	15.2%	28.5%
Within-group	100.3%	100.4%	98.5%	98.0%	90.3%
Covariance	2.9%	1.1%	-10.5%	-13.2%	-18.8%
Total	100%	100%	100%	100%	100%

What is behind the decline?

Due to the secular decline in job-separation?

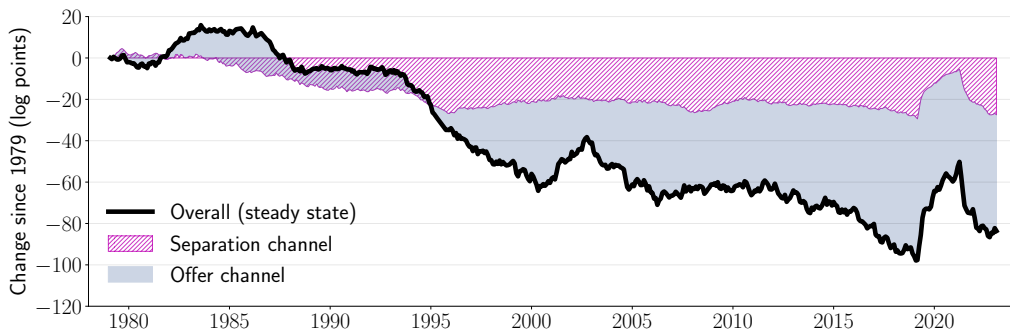
- Higher separation means workers must re-start job ladder climb more often



Due to the secular decline in job-separation?

- Higher separation means workers must re-start job ladder climb more often
- In steady state we have

$$EE = \underbrace{\delta_t}_{\text{separation channel}} \times \underbrace{\int_{-\infty}^{\infty} \frac{F_{t+1}^n(w) - G_t(w)}{G_t(w)} dw}_{\text{offer channel}}$$



Testing 3 hypotheses

Consider 3 hypotheses consistent with a decline in EE mobility.

1. Better matched workers
2. Worse matching efficiency
3. Higher firm labor market concentration

Better matched workers?

- Did EE mobility fall because workers are better matched today?

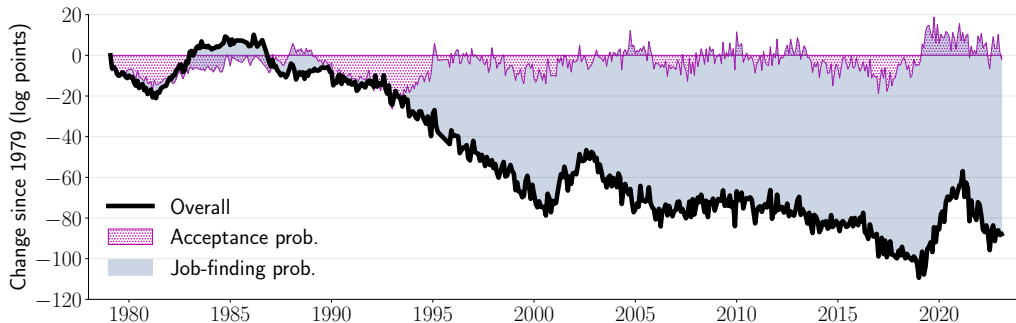
Better matched workers?

- Did EE mobility fall because workers are better matched today?

- Recall $EE_t = \underbrace{\lambda_t^e}_{\text{job finding prob.}} \times \underbrace{\int \left(1 - F_{t+1}^e(w)\right) dG_t(w)}_{\text{acceptance prob.}}$

Better matched workers?

- Did EE mobility fall because workers are better matched today?
- Recall $EE_t = \underbrace{\lambda_t^e}_{\text{job finding prob.}} \times \underbrace{\int \left(1 - F_{t+1}^e(w)\right) dG_t(w)}_{\text{acceptance prob.}}$
- Assuming $F^e = F^n$, all EE decline is from job-finding prob.



Worse matching efficiency

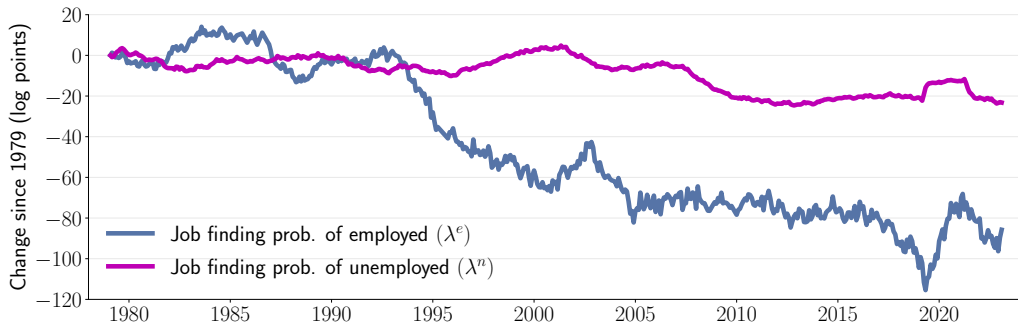
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Worse matching efficiency

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- In DMP models $\lambda^e = s \cdot \lambda^n \implies \lambda^e$ and λ^n should have **fallen proportionately**

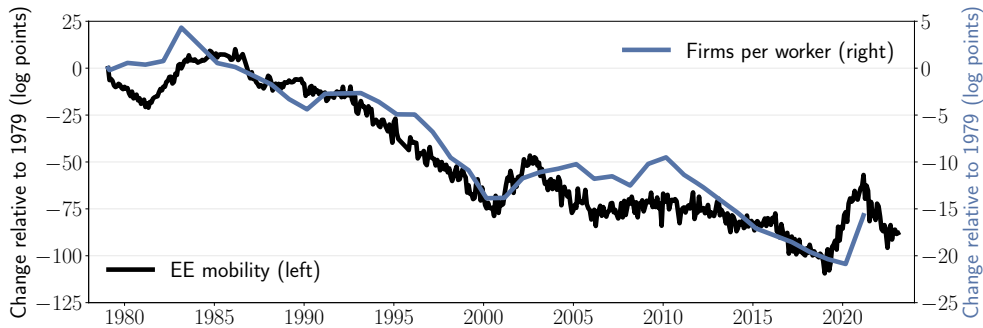
Worse matching efficiency

- Did EE mobility fall because the matching efficiency is worse?
- In DMP models $\lambda^e = s \cdot \lambda^n \implies \lambda^e$ and λ^n should have fallen proportionately
- Theory inconsistent: large decline in λ^e with little decline in λ^n



Higher firm labor market concentration

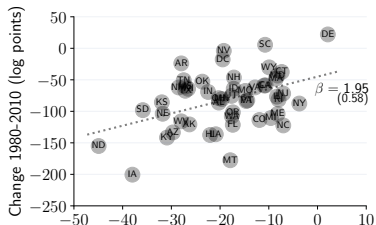
- Did EE mobility fall because of **increased firm market concentration**?
 - * Higher market concentration lowers workers' opportunities to switch employers



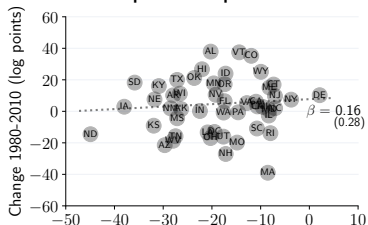
Within-state changes: 1980s - 2010s

► Counterfactual

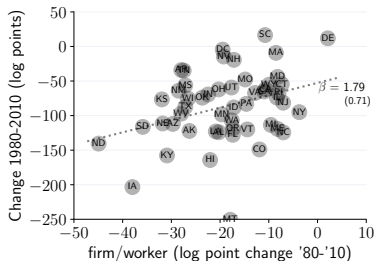
EE



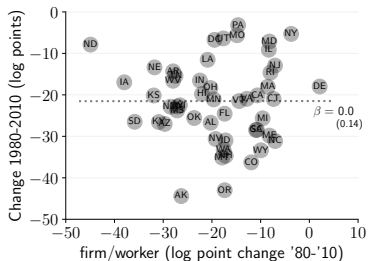
Acceptance prob.



λ^e



λ^n



Conclusion

- Estimate EE mobility halved since 1979 using job-ladder model and public data
 - * As a consequence, associated annual wage growth fell by over 1 p.p.
 - * Bigger declines for women, non-college educated workers, and newer cohorts
- Framework suggests EE decline:
 - * inconsistent with better matches and worse matching efficiency
 - * consistent with increased firm market concentration

Appendix

OTJ wage growth

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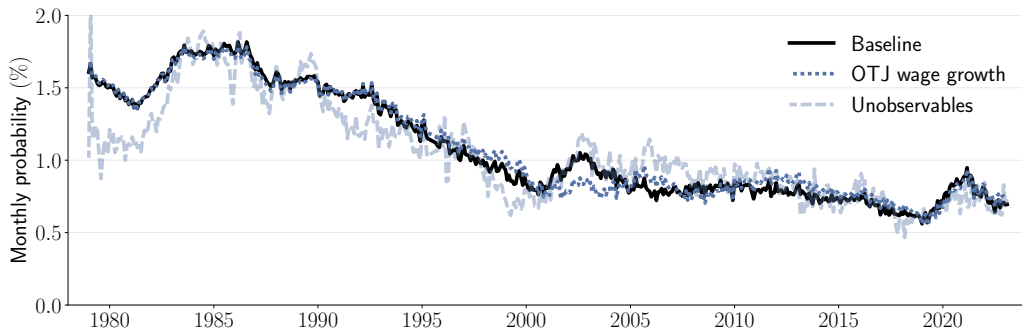
- With OTJ wage growth, EE expression is

$$\lambda_t^e \left(1 - F_{t+1}^e(w) \right) = 1 - \frac{G_{t+1}(w)}{G_t(w)} \frac{e_{t+1}}{e_t} + \lambda_t^n \frac{F_{t+1}^n(w)}{G_t(w)} \frac{1 - e_t}{e_t} - \delta_t - \frac{\tilde{\zeta}_t g_t(w)}{G_t(w)}$$

- ζ_t is the rate at which log residual wages grow on-the-job
- Estimated using the CPS *Tenure Supplement*

EE mobility: alternative specifications

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Wage growth derivation

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$$\Delta w_t^{EE} = \lambda_t^e \int_{-\infty}^{\infty} \int_w^{\infty} (\tilde{w} - w) dF_{t+1}^e(\tilde{w}) dG_t(w) = \lambda_t^e \int_{-\infty}^{\infty} \int_{-\infty}^w (w - \tilde{w}) dG_t(\tilde{w}) dF_{t+1}^e(w)$$

Integrating first the inner integral by parts

$$\Delta w_t^{EE} = \lambda_t^e \int_{-\infty}^{\infty} \left(\left[(w - \tilde{w}) G_t(\tilde{w}) \right]_{-\infty}^w + \int_{-\infty}^w G_t(\tilde{w}) d\tilde{w} \right) dF_{t+1}^e(w) = \lambda_t^e \int_{-\infty}^{\infty} \int_{-\infty}^w G_t(\tilde{w}) d\tilde{w} dF_{t+1}^e(w)$$

Integrating the outer integral by parts

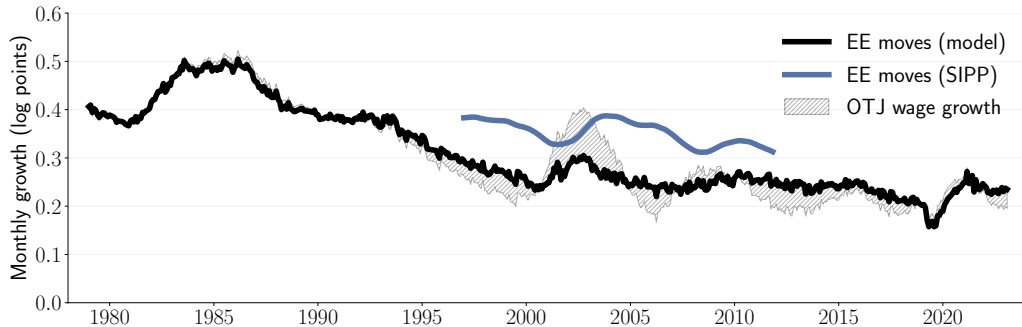
$$\Delta w_t^{EE} = \lambda_t^e \left(\left[\int_{-\infty}^w G_t(\tilde{w}) d\tilde{w} F_{t+1}^e(w) \right]_{w=-\infty}^{\infty} - \int_{-\infty}^{\infty} G_t(w) F_{t+1}^e(w) dw \right)$$

Since $\lim_{w \rightarrow \infty} F_{t+1}^e(w) = 1$, we have the following

$$\Delta w_t^{EE} = \lambda_t^e \int_{-\infty}^{\infty} (1 - F_{t+1}^e(w)) G_t(w) dw$$

Wage growth OTJ

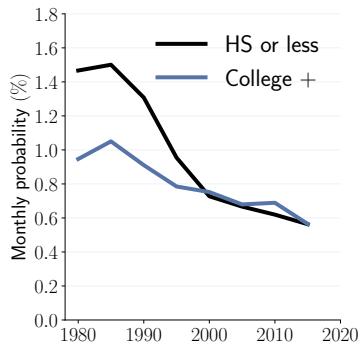
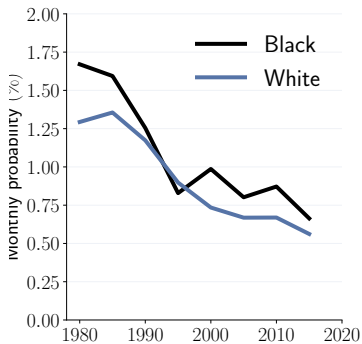
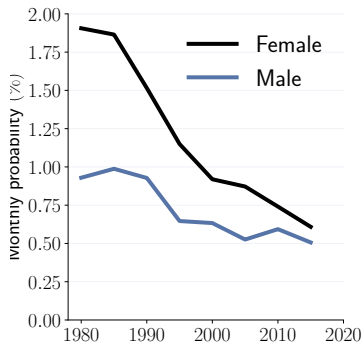
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- On the job wage growth accounts for little growth in residual wages

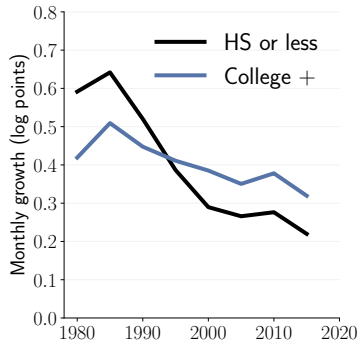
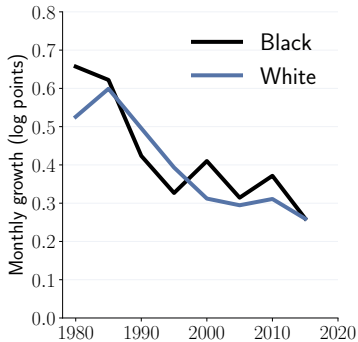
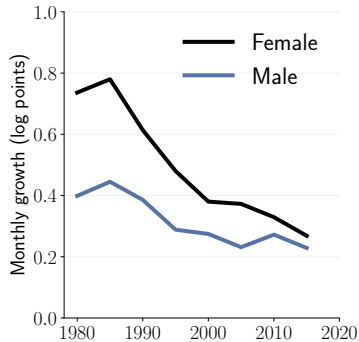
EE by race

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Wages by sub-groups

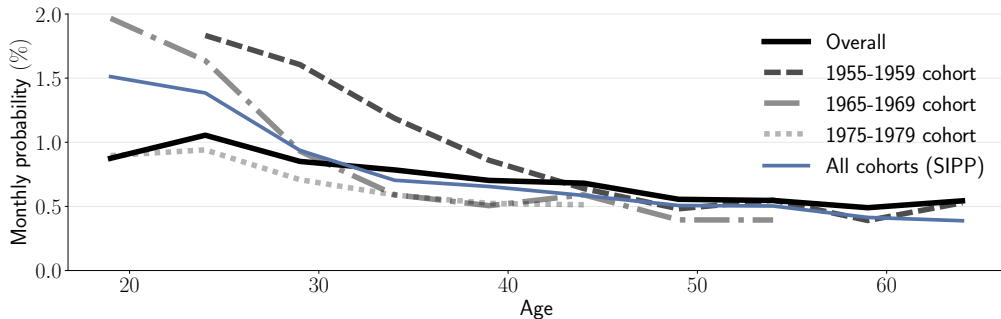
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Age, cohort, or time?

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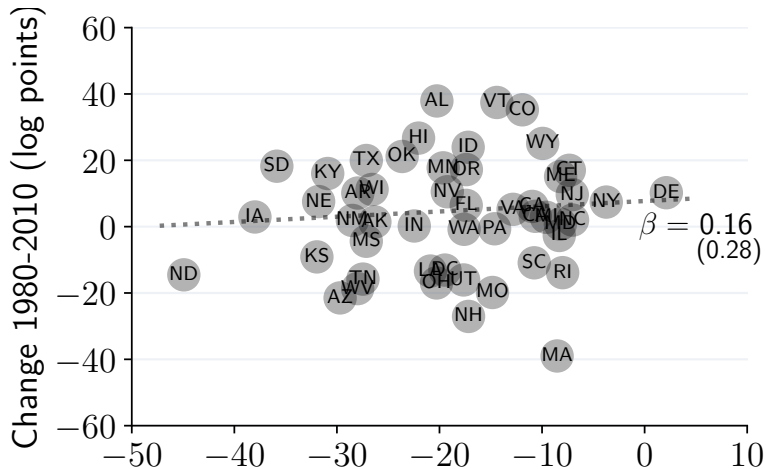
- Can quantify **role of cohorts** (Heckman et al., '98)
 - * $EE_{t,a,c} = \phi_t + \psi_a + \xi_c + \varepsilon_{t,a,c}$
 - * leverage fact that mobility is constant for older workers



Identifying assumption: No age variation in EE rate for older workers

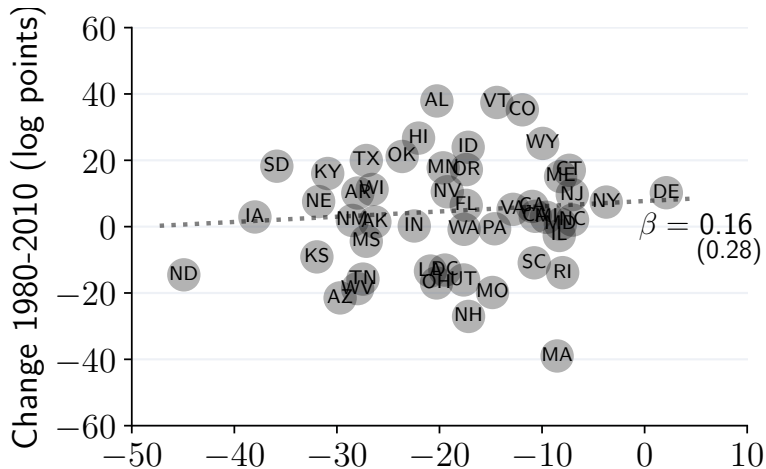
Acceptance rate by concentration

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Acceptance rate by concentration

► Back



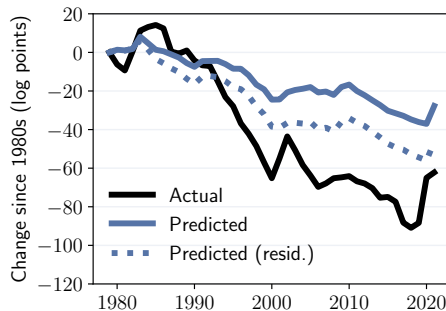
Counterfactual

► Back

- EE to concentration elasticity

$$EE_{s,t} = \beta c_{s,t} + \zeta_s + \phi_t + \varepsilon_{s,t}$$

- $\beta = 1.793$ with SE 0.556
- $1\% \uparrow \text{firms/worker} \Rightarrow 1.8\% \uparrow \text{EE}$



- Concentration increase can account for 40% of decline